Informedia
Summarization-on-Demand

Auto-Summarization and Visualization Over Multiple Documents and Libraries

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Outline

- Visualization & Summarization Goals

- Visage Dynamic Query Histograms

- Informedia Result Set Visualization (Demo)

- Result Set Clustering and Labeling

- Summarization through Title Generation

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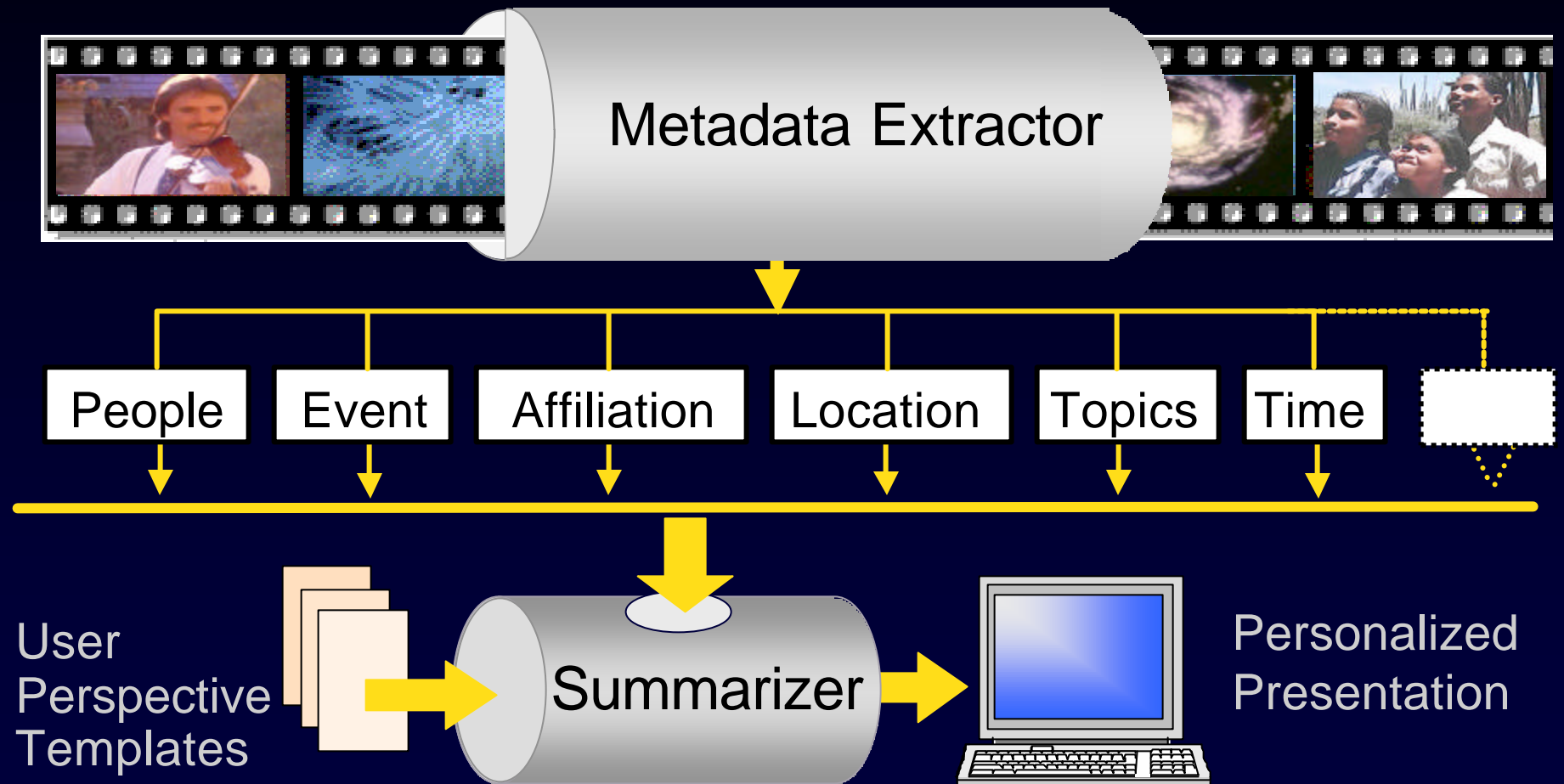# Summarization and Visualization Over Multiple Documents

Goal: Auto-summarization and visualization of multiple audio & video documents, stories and reports

- *collages*: by time, location, personalities, events
- eliminating redundancy
- at any level of detail
- creating personality and event profiles on-demand

Requisite subgoals:

- establishing *geographic* and *temporal context*
  - geocoding all data by source and internal reference
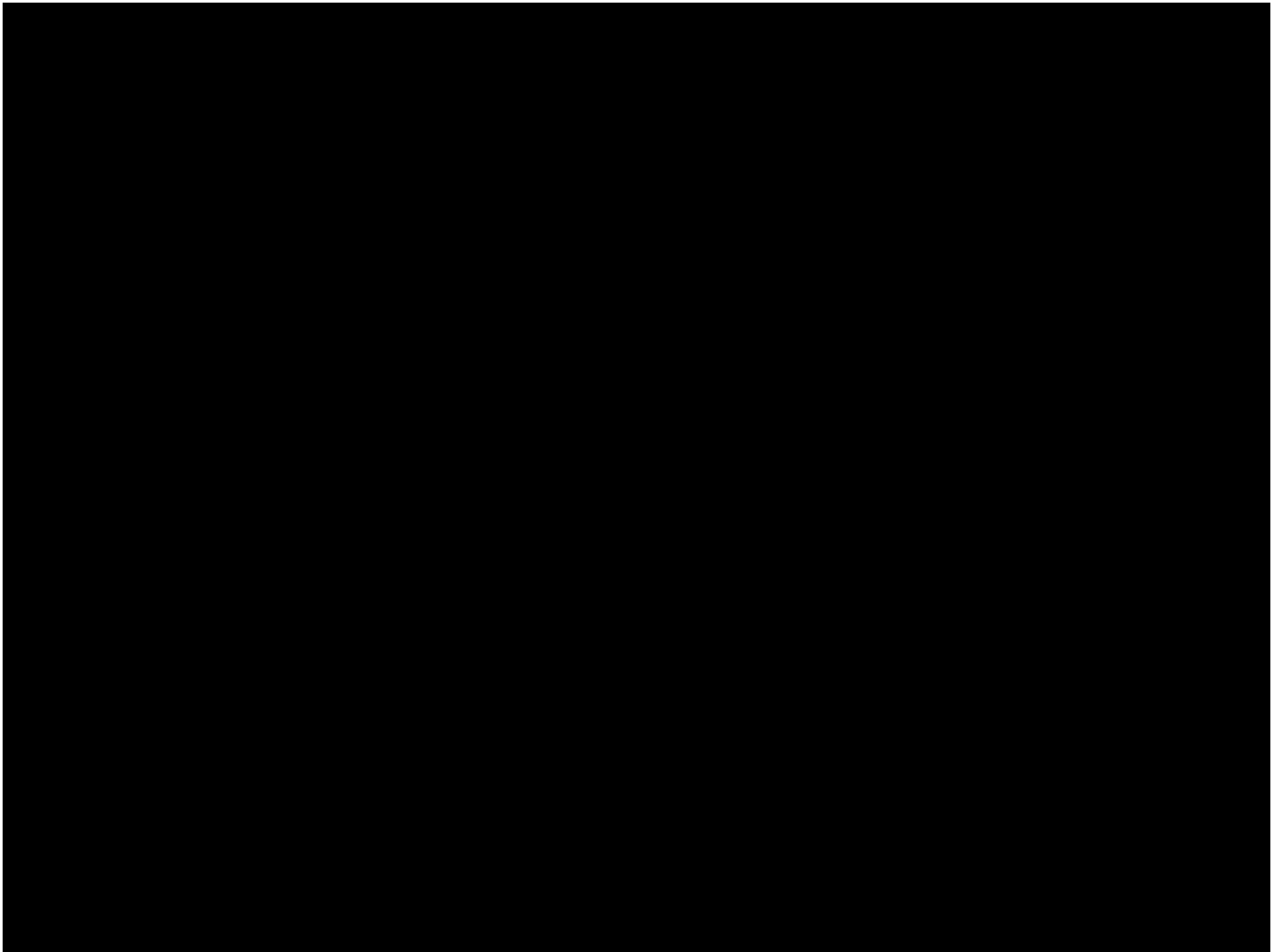 - extracting explicit and indirect time/date/place references

Metadata for Video Summarization
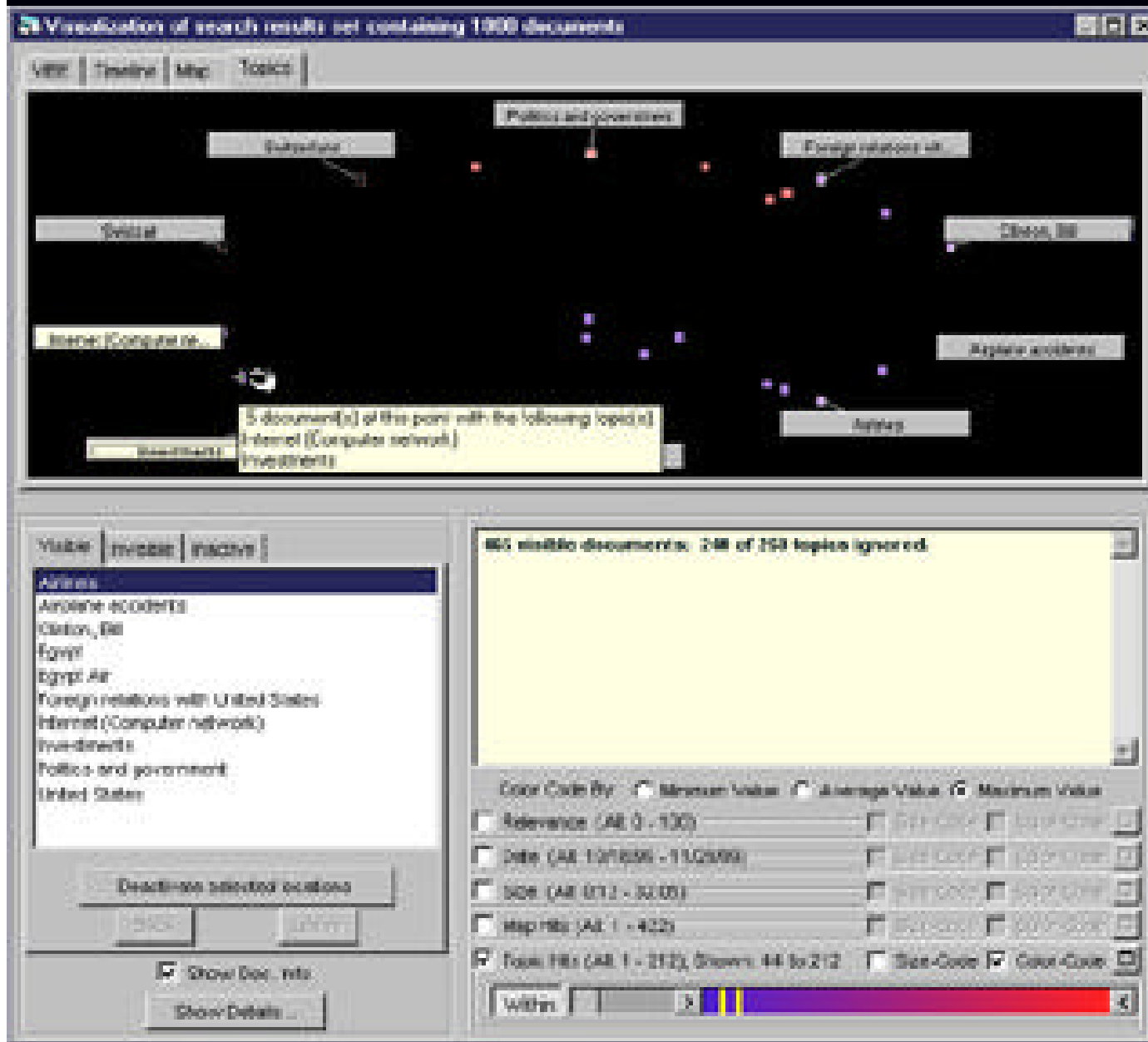
Dynamic Query Histograms
Mark A. Derthick

- Informedia library information through VISAGE data visualization tool

- Dynamically linking/updating different views on the same objects

- Efficient for large amounts of data
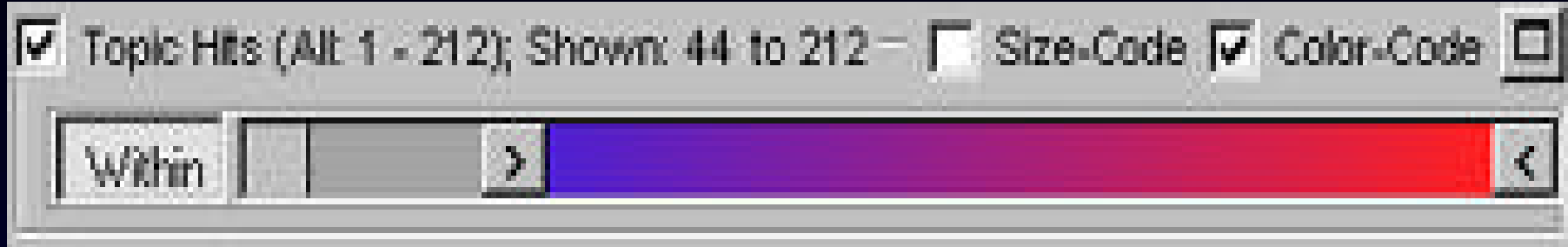
DLI-1: KNN-based Topic Detection

- Build training index with pre-labeled topics
  - 45000 Broadcast News stories from 1995 and 1996
  - 3178 different news topics occurring > 10 times

- Search for top 10 related stories in training index

- Lookup topics for related stories

- Re-weight topics by story relevance (select top 5)

- At 5 topics,   Recall - .491   Relevance - .482
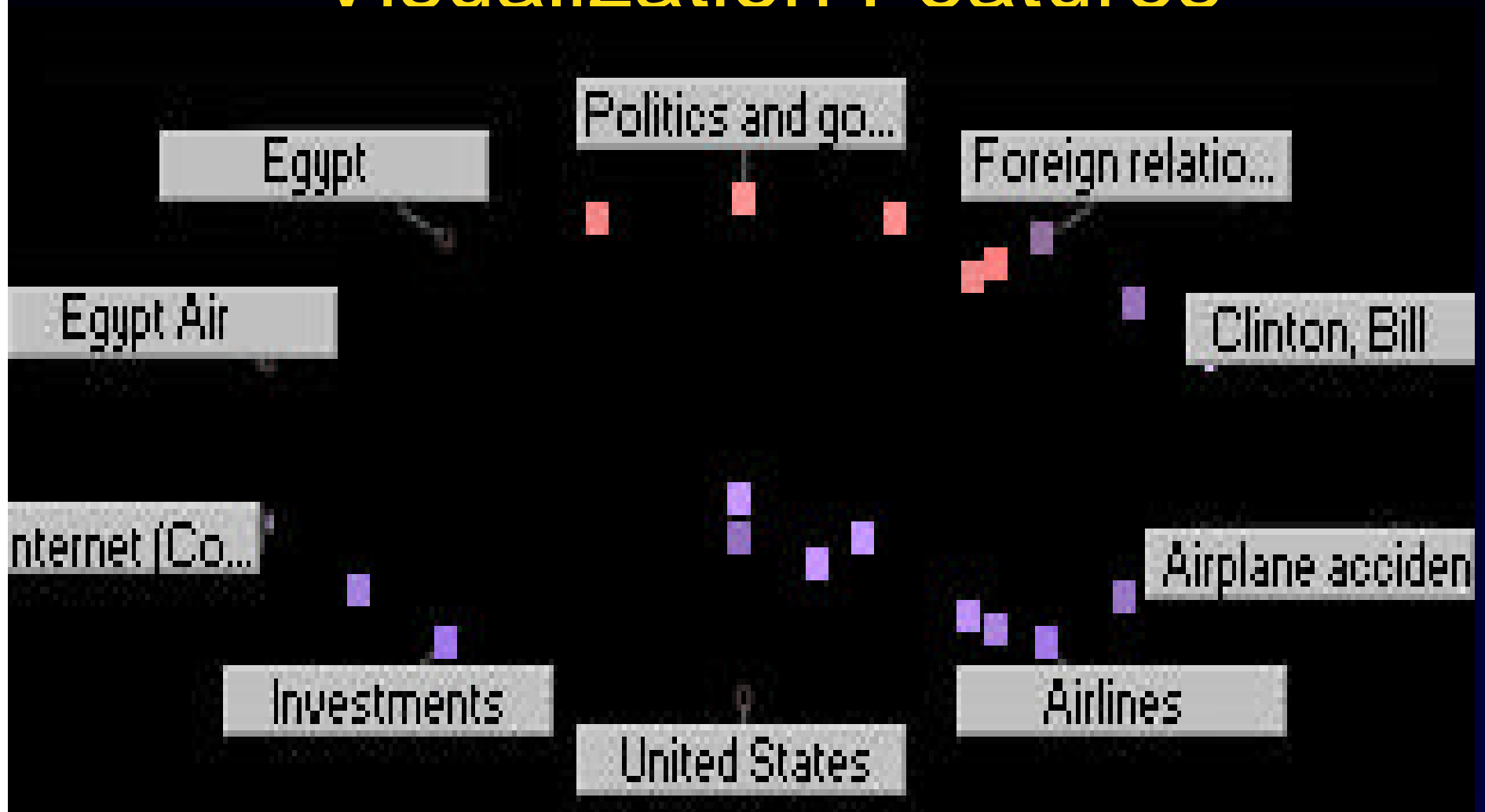
Visualization of Result Space

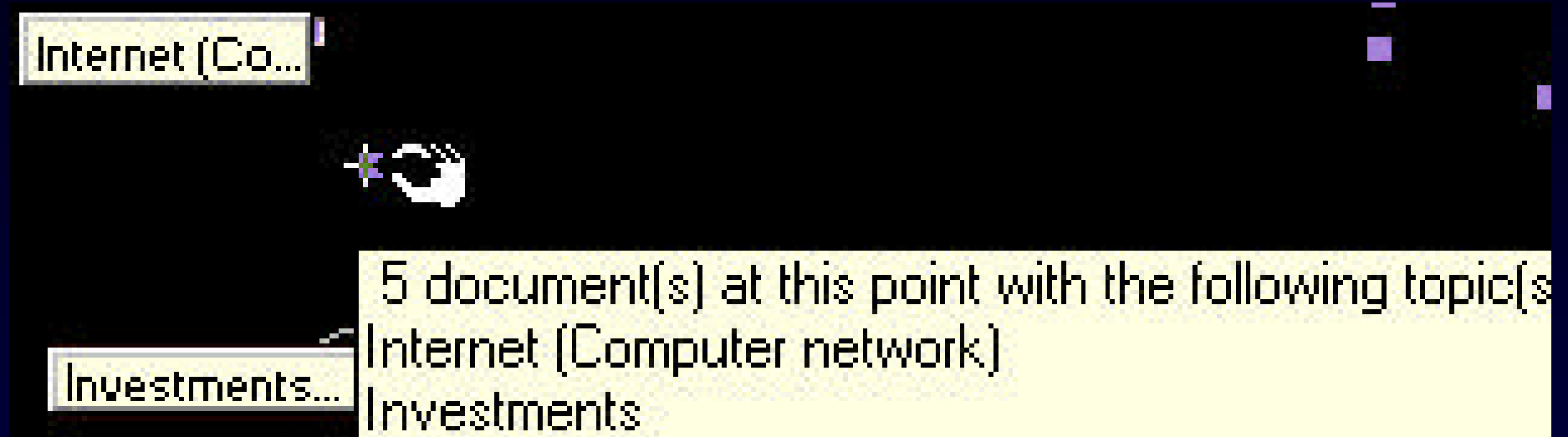Topic digest for recent news, color-coded and limited to frequently occurring topics

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Visualization Features

Slider to reduce topic digest to frequently occurring topics

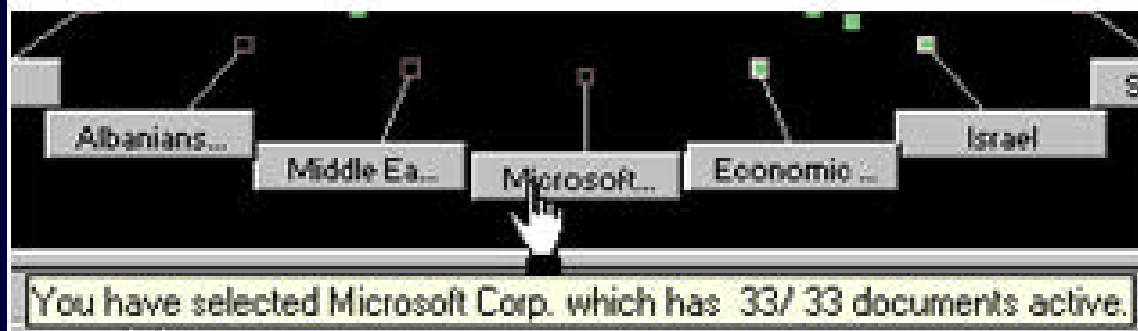# Visualization Features

Egypt

Politics and go...

Foreign relatio...

Clinton, Bill

Egypt Air

Internet (Co...

Airplane acciden

Investments

United States

Airlines

Red color shows frequent topics and blue less frequent ones

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Visualization Features

Internet (Co...

Investments...

5 document(s) at this point with the following topic(s
Internet (Computer network)
Investments

List of Active Topics

Visible	Invisible	Inactive

Airlines
Airplane accidents
Clinton, Bill
Egypt
Egypt Air
Foreign relations with United States
Internet (Computer network)
Investments
Politics and government
United States

Zooming into documents with "Microsoft Corp." as a topic

You have selected Microsoft Corp. which has 33/33 documents active.

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Visualization Features

Plot area after user zooms to documents labeled with "Microsoft Corp." topic

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Map Digest View
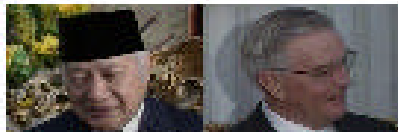

color-coded to broadcast date (blue is older, red is newer) showing distribution of documents with "guided missiles" topic

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Visualization Features

View into small set of documents, showing thumbnail images and one title
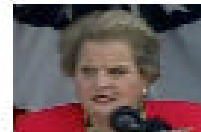
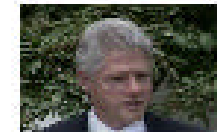# Next Step:
Topic/Face/Chrono-Collage

March 1998   April 1998   May 1998

Suharto economic reform meetings

U.S. policy on Indonesia

Habibie new president

El Niño wildfires

Student protests against Suharto

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Title Generation for Informedia News Stories

- Informedia, a multimedia digital library, stores television broadcast news stories.

- An extractive summary feature currently locates snippets in news-story transcripts to use as story titles.

- GOAL: An improved, non-extractive title-generation feature for Informedia.

Basic Idea

- Train a statistical model on a corpus of documents with human-assigned titles.
- Compare 3 title generation methods:
  - Extractive Titles
  - Naïve Bayes
  - KNN
- Apply to machine translated documents

$$F1 = \frac{(2 * precision * recall)}{(precision + recall)}$$

- Only measured word selection, not order

Experiment

- 40000 TV news stories with titles from 1998 Broadcast News CD-ROM

- tested on 10000 held-out stories evaluated on titles

Extractive Summarization

- MS Word 2000 AutoSummarize
- Extracts sentences/fragments as summaries
- Similar performance to TF IDF implementation at CMU
- Does not use our training corpus
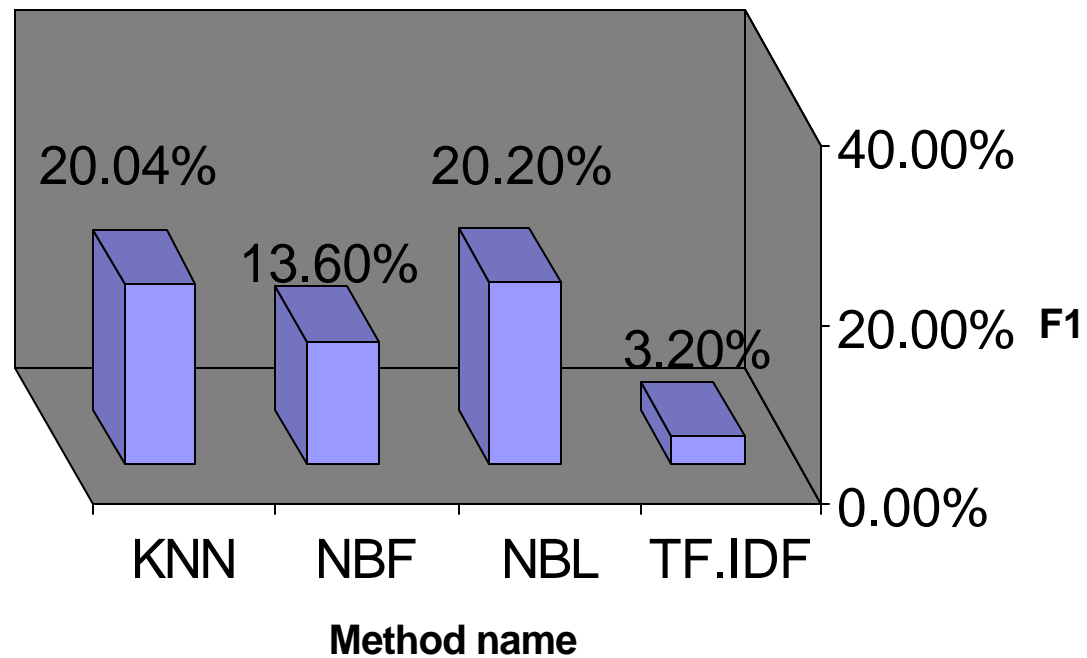
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Naïve Bayes

- Train a statistical model on a corpus of documents with human-assigned titles.

- Title need not be a snippet from the document (contrasts with extractive-summarization techniques).

- Suggested by Witbrock & Mittal, 1999.

- $P(w_{Title}|w_{Doc})$
  - works better if $W_{title} = W_{Doc}$
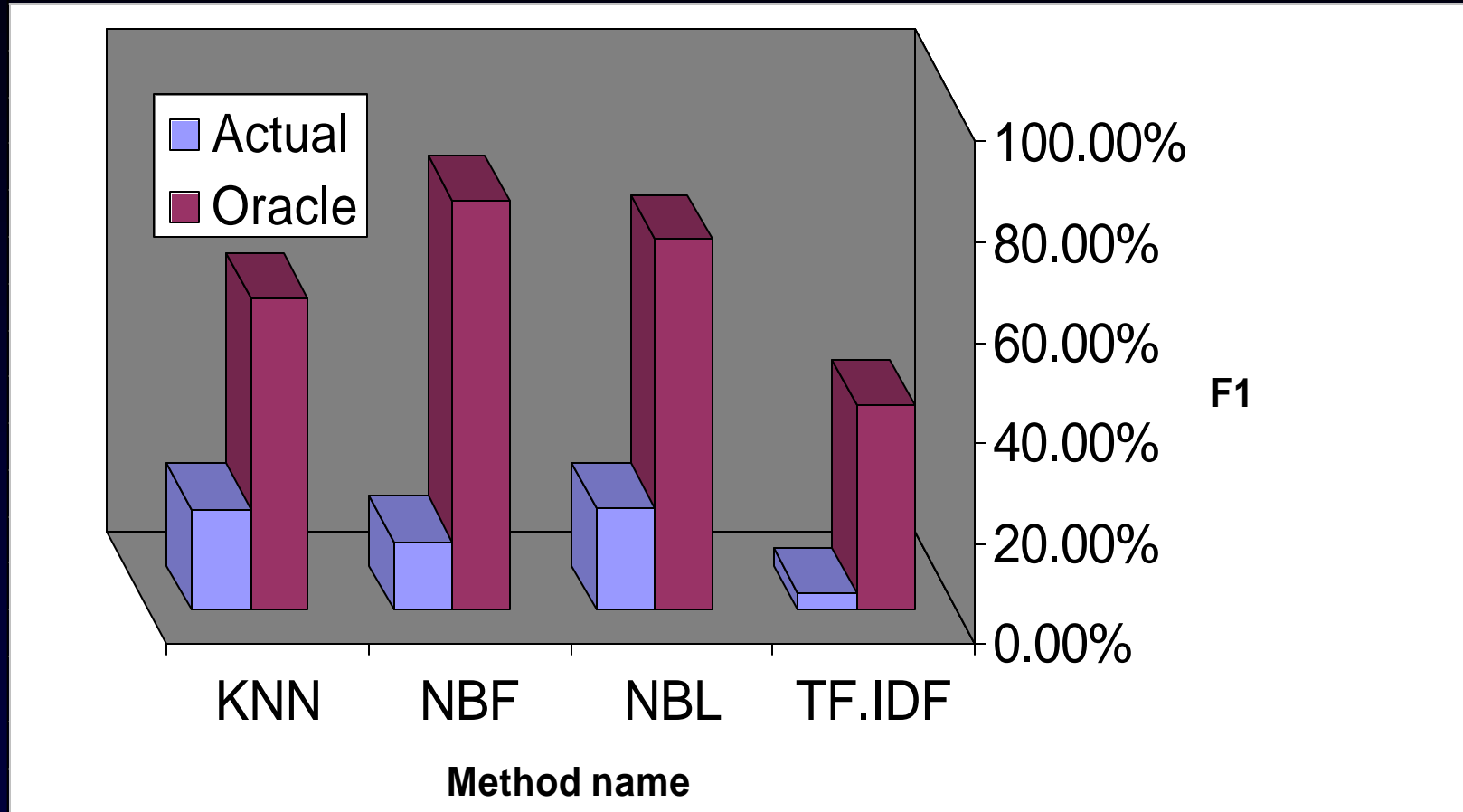
(K) Nearest Neighbor

- Index a corpus of documents with human-assigned titles.
- Find the document in the training corpus closest to the current document
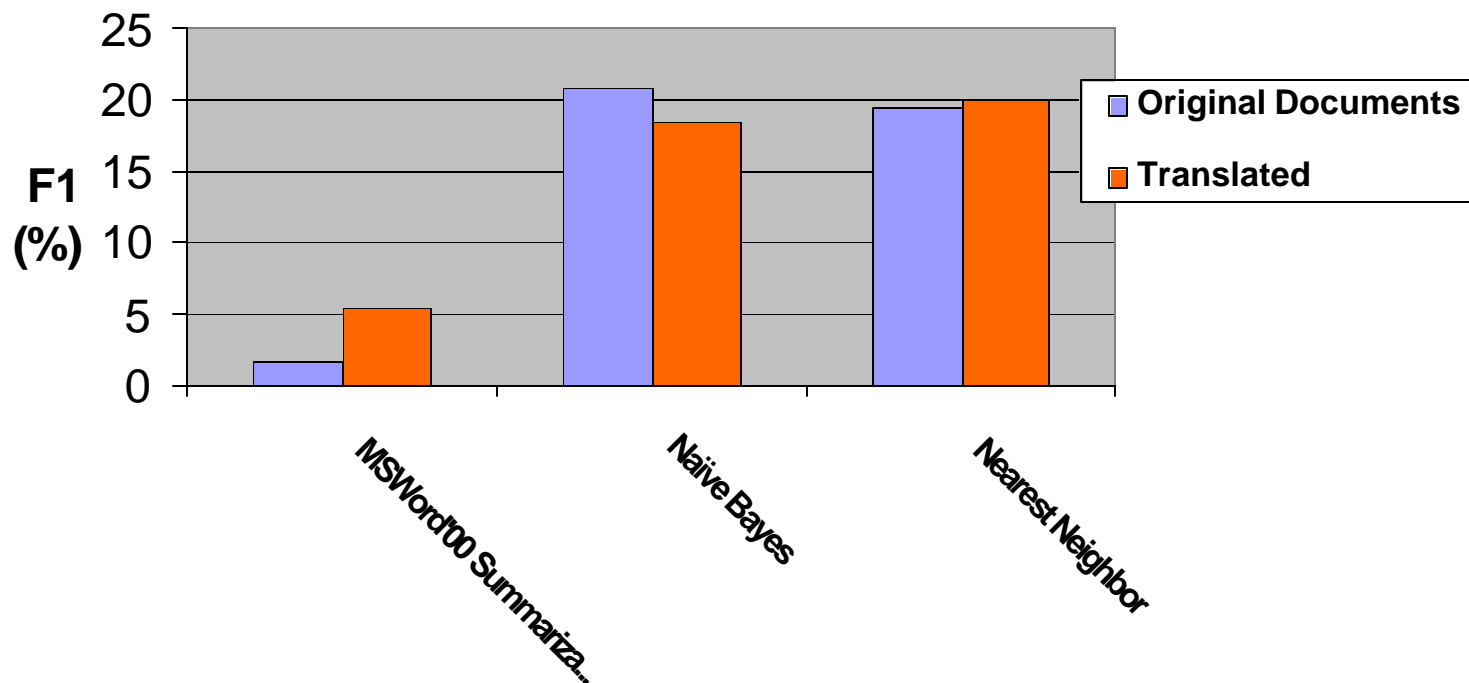- Use that title (k=1)

Results

Simulated Best Possible Results

Translation Results

Title Word Accuracy for Title Generation from **200** **English-French-English SYSTRAN Translated Documents**

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Orig Title: *INTERVIEW WITH CHINESE DISSIDENT WEI JINGSHENG*

KNN Title: *COMPLETE INTERVIEW WITH WEI JINGSHENG*

MSWORD AUTOSUM TITLE: *IT WAS A DECISION TO COME TO THE UNITED STATES ON BEHALF OF THE CHINESE GOVERNMENT STEP A DECISION TO SEEK THE GOOD PROMPT MEDICAL PROCESSING FOR YOU.*

Naïve Bayesian Title: *china wei chinese jingsheng china's prison*

SYSTRAN TRANSLATED TEXT (TO FRENCH AND BACK TO ENGLISH):

- IN AN EXCLUSIVE INTERVIEW WITH CHINESE DISSIDENT WEI OF C N N EXILED JINGSHENG INDICATES THAT IT SUPPORTS THE CONFRONTATION OF DIALOGUE NOT WITH THE CHINESE GOVERNMENT. RALPH BEGLEITER OF CORRESPONDENT OF BUSINESSES SAID BY WEI OF THE WORLD OF C N N HOW IT COULD COME TO THE UNITED STATES AND WHAT IT A DUE TO SUPPORT DURING ITS YEARS IN A CHINESE PRISON. IT WAS A DECISION TO COME TO THE UNITED STATES ON BEHALF OF THE CHINESE GOVERNMENT STEP A DECISION TO SEEK THE GOOD PROMPT MEDICAL PROCESSING FOR YOU. THEY CLEARLY INDICATED TO ME THAT IT IS IMPOSSIBLE SO THAT I REMAIN IN CHINA. IF I WANT TO OBTAIN A MEDICAL PROCESSING THAT YOU MUST GO ABROAD AND YOU MUST GO TO THE UNITED STATES. IT IS A CONDITION WHICH THEY CLEARLY INDICATED TO ME THUS. SAY A LITTLE TO ME ABOUT THE TRUTH OF YOUR PROCESSING IN PRISON. WERE YOU TORTURED WERE YOU WOUNDED BY THE GOVERNMENT OR OTHER PRISONERS. YES. USUALLY THESE BEATS ARE NOT DIRECTLY OF THE FONT. THEY USE USUALLY CRIMINALS WITH BEATUP OF OTHER CRIMINALS. IT BEATS ME NOT ONLY THAT IT IS COMMON OCCURRENCE WHICH TORTURES AND WHILE BEATING PRODUCE YOU IN THE PRISON. WEI INDICATES THAT IT SE RETURNS ACCOUNT WHICH IT THERE A OF A MORE ECONOMIC FREEDOM IN CHINA THAN THERE WAS BUT IT SAYS THAT POLITICAL FREEDOM IS MORE SIGNIFICANT AND THAT ONE REFUSED TO HIM WITH

  THE PEOPLE OF CHINA

Title Generation Conclusions

- Training on a corpus really helps
- But coverage of training to testing better be good
- KNN is preferred since it provides a good readable title
- KNN could be improved with K > 1,
  - but then we need to combine title fragments
- Title word learning can also be improved
  - but we need robust word serialization with a language model

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Other Insights

- Similar performance achieved using news stories with titles harvested from the web

- EM instead of Naïve Bayes is slightly better

- Small degradation for corrupted text (speech recognition, machine translation)

Summary of Summaries and Visualizations

- Exploring different paradigms of visualization
  - No 'Right Way'

- Summarization through labeling and titles

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The End

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